# Face Show

## documentation Sept.-Oct. 2023

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a chest-up selfie photo of prkrdvs person holding bread in the store, tree in background



a chest-up selfie photo of prkrdvs person wearing a black shirt, holding bread in the store, tree in background

the angle and quality of the photo, including how much of the person you can see the token name + the word person.

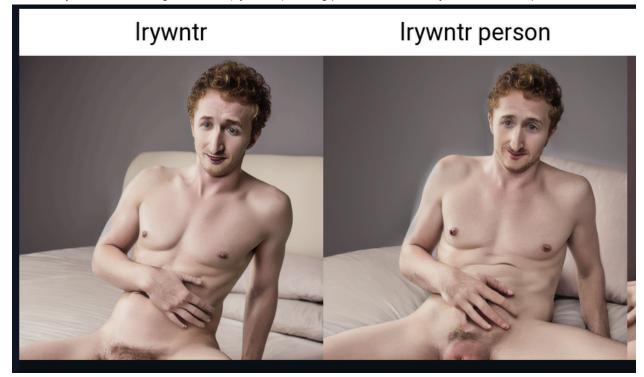
characteristics of the person, include anything that's unique about the photo and the figure which shouldn't be learned in the training (e.g. parker does not always have a black shirt and isn't always in the store). Be specific about anything in the photo. I didn't tag parker's glasses because he always has glasses on.

- 1. train ai on everyone's face
  - a. get 3-5 selfies from each RSVP'd person
  - b. train textual inversion over the week w/ each person,
  - c. take photos of people in the gallery to train it on the location
  - d. create a template for the text-to-image generations
- 2. create text to speech
  - a. make a list of conversation prompts for inserting speech-to-text input
  - b. have microphone do speech to text and record samples from the party
  - c. create a framework to send text from microphone into opensource LLM
  - d. use the transcribed conversation snippets to create text-to-image prompt and chatbot prompt at the same time
  - e. project photo outputs with text rendered below, use tts to read out the dialogue.
    - i. link photo with text output under the name of the speaker, send to one of two screens somehow
    - ii.
    - iii.
    - iv.
    - v

Process notebook

#### notes on prompts:

use the keyword and the weight like this: (Irywntr:1). adding person unnecessarily makes it more "person-like",



#### regional prompter



candid photo of 3people at a hipster party  ${\tt BREAK}$ 

(nnsnmn :1) BREAK
(gbrlmcg :1) BREAK

(drsshl :1)

Steps: 20, Sampler: DPM++ 2M Karras, CFG scale: 7, Seed: 4191126256, Size: 1024x512, Model hash: 79b5166ee6, Model: last-flippedZcond-ep400-gs05500, RP Active: True, RP Divide mode: Matrix, RP Matrix submode: Columns, RP Mask submode: Mask, RP Prompt submode: Prompt, RP Calc Mode: Attention, RP Ratios: "1,1,1", RP Base Ratios: 0.2, RP Use Base: True, RP Use Common: False, RP Use Ncommon: False, RP Change AND: False, RP LoRA Neg Te Ratios: 0, RP LoRA Neg U Ratios: 0, RP threshold: 0.1, RP LoRA Stop Step: 0, RP LoRA Hires Stop Step: 0, RP Flip: False, Version: v1.6.0

#### MULTI FACE ADETAILER

https://github.com/Bing-su/adetailer/wiki/Advanced https://github.com/Bing-su/adetailer/discussions/204 normal prompt:

#### adetailer prompt:

(nnsnmn:1), 1girl [SEP] (lxbth:1), 1girl [SEP] (gbrlmcg:1), 1boy [SEP], (drsshl:1), 1boy LORA model training

- follow github tutorial for runpod (may need diff tut for alienware
  - https://github.com/FurkanGozukara/Stable-Diffusion/blob/main/Tutorials/How-To-Install-Kohya-LoR A-Web-UI-On-RunPod.md#auto-installer-for-runpod
- download regularization images:
  - waet

https://github.com/FurkanGozukara/Stable-Diffusion/blob/main/Tutorials/How-To-Install-Kohya-LoR A-Web-UI-On-RunPod.md#auto-installer-for-runpod

- Testing 3 images of alex, 1024x1024, max res 512x512
- alextest1
  - o using sdxl produces massive checkpoints for lora, tested w/ sdxl base
  - woman at a party as class identifier
- alextest2
  - testing using all default except no xformers and network size of 256.
  - o woman as class identifier
  - o base model:

/workspace/stable-diffusion-webui/models/Stable-diffusion/Realistic Vision V5.1.safetensors

- alextest3
  - testing using all default except no xformers and network size of 8.
  - woman as class identifier
- alextest4
  - network size 8
  - woman at a party with white walls, size 8 network
- alextest5
  - network size 8
  - o repeats 20 (instead of 40)
  - o xformers on
  - o 1024x1024
  - woman in a white room class prompt
- alextest6 (20230917) alex's computer
  - same as alextest4
  - o network size 8
  - new reg images from party
- alextest7 gabe's computer
  - o new reg images
  - woman in white room"
- alextwst8 alex's computer
  - new reg images
  - o 20 epochs
  - o 60 network size
- alextest9 darius computer
  - network rank 8repeats:60
  - alaytaat10 dariya sara
  - alextest10 darius computer
    - o batch size 2
    - o constant Ir scheduler

0

Current test is with iranian dude's recommendation of a network size of 256. this may be way the lora models are big but probably makes good results. another option from <a href="reddit">reddit</a> says to make `lora\_rank` smaller

use automatic1111 dreambooth training supposedly, if no .txt is found, it reads the name of the file as the instance token.  use [filewords] in the instance token use person for class token use full body photo of [filewords] person for instance prompt use full body photo of person for class prompt profit?  **alex3.jpg alex4.jpg alex6.jpg alex6.jpg **alex6.jpg **oreinstalling microsoft malware or you will to be losing all of investment portolio to hacker sir  **alex6.jpg **alex1.jpg **alex1.jpg **alex3.jpg **alex1.jpg **alex3.jpg **alex1.jpg **alex1.jpg **alex3.jpg **alex1.jpg **alex3.jpg **alex1.jpg **alex3.jpg **alex1.jpg **alex3.jpg **alexdarius test 2 **make a txt files containing a unique instance token(1 per subject) for adarius5.jpg **alexdarius test 2 **make a txt file corresponding to each photo like this picture ***alex3.jpg **alex3.jpg **alex4.jpg **alex6.jpg	NEW TECHNIQUE		Name	
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	multi-subject lora tests on Kohya	⊗ Ø dar	ius3.txt	
repeats dont matter as long as all are around 100, (its for balancing multiple drsshl	repeats dont matter as long as all are around 100, (its for balancing multiple	drsshl		
	images seperately)			
	standard lora with 32 network size is pretty good.			
defaults are pretty good repeats balancing should be used if some folders contain more images than others	defaults are pretty good	are		

#### muti-subject and place test (dreambooth realistic vision 512) test 1

captioned place photos, left subjects with 'ohxw person'

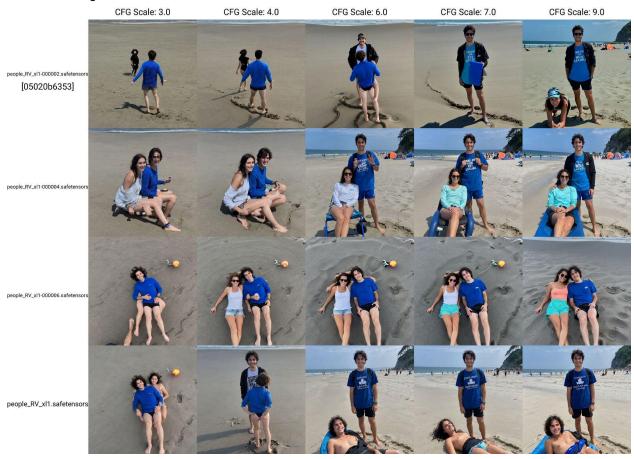
multi-subject sdxl w/ 1300 reg images

good. only works at 1024x1024. using exactly secourses json. repeats =86 per img folder faces were not learned well enough.

multi-subject Realistic vision xl 2 w/ 500 reg images

the idea: less reg images = better face memorization, lower step count. doing 33 repeats per folder. max steps 7920 results

soooo the images are all beautiful - however the faces get mixed quite easily more steps = more defined faces. This works way better than other models for some reason. created a basic grid:



## rvxl-lora big boy test settings:

```
caption_dropout_rate: 0.0
caption_dropout_every_n_epoches: 0
caption_dropout_every_n_epoches: 0
caption_tag_dropout_rate: 0.0
caption_prefix: None
caption_suffix: None
color_aug: False
filp_aug: False
filp_aug: False
filp_aug: False
token_warmup_min: 1;
token_warmup_min: 1;
token_warmup_min: 2der_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_under_un
               caption_extension:.bt

[Subset 1 of Dataset 0]
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               caption_extension:.ixt

(Subset 2 of Dataset 0)
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image_der. ("
                                                                                                                                                                                                                                                                                                                                                    udent\Desktop\kohya\kohya_ss\rvxl-lora\img\33_matvairo person"
[Dataset 0]
loading image sizes.
100%
loading image sizes.
100%
prepare dataset
preparing accelerator
loading model for process 0/1
load Stable-Dilivsion checkpoint C:\Users\student\Downloads\sd.webu\iwebu\iwebu\imodels\Stable-diffusion\ivxi-training2_RealVisXL_V2.0.safetensors
building U-Net
loading U-Net from checkpoint
U-Net - All keys matched successfully>
building text encoder from checkpoint
loading U-Net from checkpoint
loading U-Net from stable successfully>
building text encoder sizes of the stable successfully-
building text encoder 2 - All keys matched successfully>
building VAE
loading VAE from checkpoint
VAE: -All keys matched successfully>
loading VAE
loading VAE from checkpoint
AR from checkpoint
AR from checkpoint
AR from checkpoint
I-Net from 
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                      15/15 [00:00<00:00, 110.00it/s]
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           15/15 [00:00<00:00, 76.53it/s]
          100%] caching latents...
01 (00:00, 7!kis) caching latents...
01 (00:00, 7!kis) careate LoRA network, base dim (rank): 32, alpha: 1.0
neuron dropout: p=None, rank dropout: p=None, module dropout: p=None
create LoRA for Text Encoder 2:
create LoRA for Text Encoder: 284 modules.
create LoRA for U-Net: 722 modules.
enable LoRA for U-Net: 722 modules.
enable LoRA for U-Net: recorder
enable LoRA for U-Net: recorder
enable LoRA for U-Net: recorder
          mable LORA for LI-Net
prepare optimizer ($cale_parameter: False, "relative_step": False, warmup_init". False)
prepare optimizer ($cale_parameter: False, "relative_step": False, warmup_init". False)
because max_grad_nomn is set_lot_parad_nomn is enabled. consider set to 0 / max_grad_nomn/设定されているためclip_grad_nomn/有効になります。0に設定して無効にしたほうがいいかもしれ
constant_with_warmup will be good / スケジューラはconstant_with_warmupが良いかもしれません
constant_with_warmup will be good / スケジューラはconstant_with_warmupが良いかもしれません
running training / 宇宙開発
running training / 宇宙開発の数×繰り返し回数: 495
num reg images / 正則に画像の数・繰り返し回数: 495
num reg images / 正則に画像の数・繰り返し回数: 495
num balches per spoot. / 1 spoot.かりナ数: 495
balch size per device / パッチサイズ: 1
gradient accumulation steps / 須配在会計するステップ数 = 1
total optimization steps / 宇宙ステップ数: 495
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                           CFG Scale: 5.0
                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                                  CFG Scale: 7.0
```

#### results boii

random TI test (probs retarded) following this guys ideas https://www.reddit.com/r/StableD

#### **EveryDream2 testing**

test1. testing using 5 photos eacl = 5. next test i will try custom bat

test 2. batch size 1 with custom  $t_{\mbox{\tiny clora:multi3-00000}}$ batch\_size=5, so i changed it to

- result:
  - underfit in com 0
  - Batch size of 1
  - custom batchir lora:multi3-000005 0

test3. gonna continue to try custo flipped images

```
**run train.py --config train.j
--resume_ckpt "SG161222/Realis
--project_name "flipped-custom
--data_root "input" \
                                                                lora:multi3-000009
 --max_epochs 400 \
--sample_steps 150 \
--save_every_n_epochs 100 \
--lr 1.2e-6 \
--lr_scheduler constant \
--save_full_precision \
--batch_size 5
```

#### result:

- underfit even a 0
- custom batchir 0

good group photo including (1g

#### test 4: lower cond dropout, no cu

```
%run train.py --config train.jsv.. \
--resume_ckpt "SG161222/Realistic_Vision_V5.1_noVAE" \
--project_name "flipped2cond" \
--data_root "input" \
--max epochs 400 \
--sample_steps 150 \
--save_every_n_epochs 100 \
--lr 1.2e-6 \
--lr_scheduler constant \
--save_full_precision \
--batch_size 5 \
--cond_dropout 0.02
```

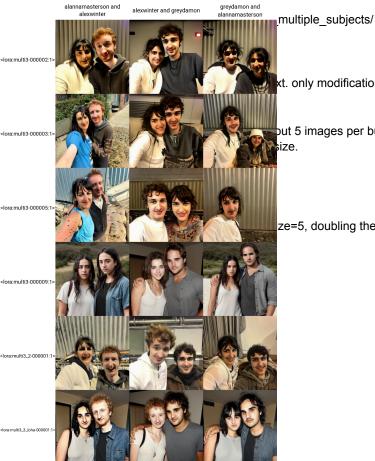
test5 same but w/ 0.07 balancing in the gt folder using multiply.txt, and batch size of 2

```
%run train.py --config train.json \
--resume_ckpt "$G161222/Realistic Vision_V5.1_noVAE" \
--project_name "flipped2condbalanced" \
--data_root "input" \
--max epochs 400 \
--sample_steps 150 \
--save_every_n_epochs 100 \
--lr 1.2e-6 \
--lr_scheduler_constant \
--save_full_precision \setminus
--batch size 2 \
```

#### Test 4 vs test 5

--cond\_dropout 0.02

verdict: test 4 @ 400epochs is best.



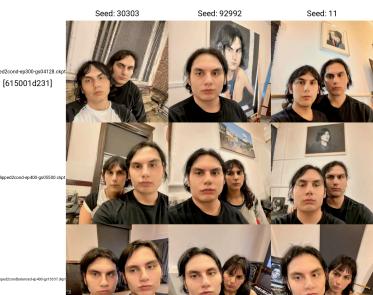
alexwinter and greydamon

alannamasterson and

kt. only modification is batch size

out 5 images per bucket with

ze=5, doubling the dataset w/



Seed: 13123 Seed: 232

## current best settings

normal prompt including multiple characters, hires fix, then use adetailer with [sep] tags in adetailer prompt, mediapipe\_full\_face.pt,

#### main prompt

amazing photo of 4people at the beach, (nnsnmn:1.2) 1girl, and (lxbth:1.2) 1girl, and (gbrlmcg:1.2) 1boy, and (drsshl:1.2) 1boy

#### adetailer prompt

(nnsnmn:1.2) 1girl [SEP] (lxbth:1.2) 1girl [SEP] (gbrlmcg:1.2) 1boy [SEP], (drsshl:1.2) 1boy





flipped-custom-batch-ep300-gs04109.ckpt

[22a4abe3db]



[15511cf88b]



last-noval-ep250-gs02236.ckpt



SDXL dreambooth 2 epoch 15 might be best, lowest loss around there using new patreon settings, no reg IT SUCKS

**back to everydream2** (tentative idea to chain it in into a higher resolution at more steps?) added 500 more GT images, using multiply.txt of 0.02 (0.02\*700=14 per batch).

```
%run train.py --config train.json \
    --resume_ckpt "SG161222/Realistic_Vision_V5.1_noVAE" \
    --project_name "flipped2cond" \
    --data_root "input" \
    --max_epochs 400 \
    --sample_steps 150 \
    --save_every_n_epochs 25 \
    --lr 1.5e-6 \
    --lr_scheduler constant \
    --save_full_precision \
    --batch_size 5 \
    --cond_dropout 0.02

^ decent results; will prob be good enough w/ adetailer.
```

#### testing at 764 w/ a lower learning rate

```
%run train.py --config train.json \
--resume_ckpt "SG161222/Realistic_Vision_V5.1_noVAE" \
--project_name "flippedcond764" \
--data_root "input" \
--max_epochs 400 \
--sample_steps 150 \
--save_every_n_epochs 50 \
--lr 0.8e-6 \
--lr_scheduler constant \
--save_full_precision \
--batch_size 5 \
--cond_dropout 0.02 \
--resolution 768
```

oct 17 running main model + batched sulkroom using batch\_id only in sulk. 200 gt images

```
%run train.py --config train.json \
--resume_ckpt "SG161222/Realistic_Vision_V5.1_noVAE" \
--project_name "flipped2cond" \
--data_root "input" \
--max_epochs 350 \
--sample_steps 150 \
--save_every_n_epochs 50 \
--lr_15e-6 \
--lr_scheduler constant \
--save_full_precision \
```

--batch\_size 10 \
--cond\_dropout 0.02

replicate / vicuna-13b ©

⊕ PUBLIC A large language model that's been fine-tuned on ChatGPT interactions

\$\mathscr{A}\$ 176.2K runs \$\mathscr{A}\$ GitHub \$\mathscr{B}\$ Paper \$\mathscr{B}\$ License

Demo \$\mathscr{A}\$ API \$\mathscr{B}\$ Examples \$\mathscr{D}\$ Versions (6282abe6)

#### Input

#### T prompt

subtitles for alex both starting with: "thats literally so dumb why would you say that"

Prompt to send to Llama.

#### # max\_length

500

SYSTEM

Maximum number of tokens to generate. A word is generally 2-3 tokens (minimum: 1)

#### Output

Alex: in sorry, i was trying to be funny

Alex2: its fine, comedy can be hit or miss

Alex: yeah, sometimes my jokes fall flat

Alex2: no worries, just be yourself

Alex: thanks, i appreciate that

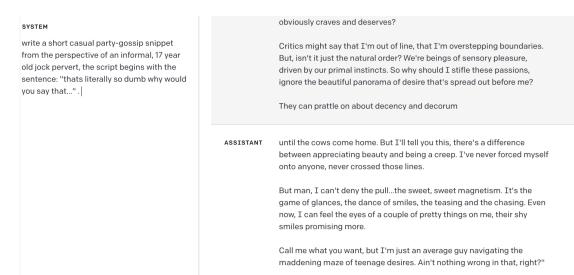
Alex2: of course, you're a great person

Alex: aw, thanks

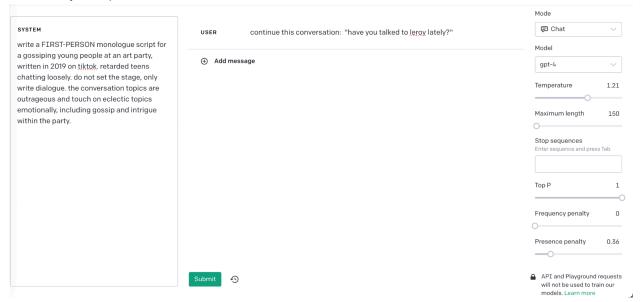
#### # temperature

write a monologue script for a gossiping college student at a party trying to get attention, written in 2019 on <u>tiktok</u> Generated in 6.30 seconds

USER	write a short snippet of dialogue spoken by a ridiculous college student at an art school party. it begins with: "thats literally so dumb why would you say that".	11000	
USER		Chat	
		Model	
		gpt-4	
ASSIST	ANT (Record button clicked)	Temperature	1.15
	"Oh, my gosh! That's literally so dumb, why would you even say that?	$\overline{}$	
	Okay sunscreen enthusiasts, listen up! Now Jeremy from Design 101 has	Maximum length	150
	ditched his uptight country club look and is officially dialing up his edgy quotient. He dyed his hair fuchsia last night, yes FUCH-SIAA! Exactly like	0	
	Fiona's hair! Related or suspect?"	Stop sequences Enter sequence and pres	s Tab
	(Turns camera to self with a fake dramatic gasp)		
	"And hey, did you guys know? Sarah's vertical garden that she can't stop posting about, which by the way, got 500 more likes? Bought from an	Top P	1
	online store! Ikea has suddenly beamed into our world, disrupted our noble efforts and is	Frequency penalty	0
① Add	l message	Presence penalty	0
Submit		API and Playground will not be used to tra	



### vii. Chat GPT system qualifiers:



Hey chat, I would like you to compose a FIRST-PERSON script of gossiping young people at an art party, written in 2023 on tiktok, retarded teens chatting loosely. Do not set the stage, only write dialogue, the conversation topics are outrageous and touch on eclectic topics emotionally, including gossip and intrigue within the party. Below I have created 2 main lists. The first list is of Names I would like you to use exclusively for names of the characters in the script you generate. The second list is of Chicago Neighborhoods, Venues, and Bars. You can use these locations as part of the conversation, but you DO NOT HAVE TO. Maybe 0-2 locations per script.

Names: Darius, Alex, Gabriel, Leroy, Allegra, Parker, Nell, Ava, Henry, Max B., Anna, Lucas

Neighborhoods: The loop, Gold Coast, Old Town, South Loop, Pilsen, Chinatown Lakeview, Wrigleyville, Wicker Park, Bucktown, Ukrainian Village, Bridgeport, Little village, Wigga Park, Logan, East Pilsen, River North, Gold Coast, Albany Park

Venues: The Bottle, Clipper, Podlasie, Smart bar, Archer, Laura, Sleeping Village, Cafe Moustachio, The Hideout, Thalia Hall, DADS, Co-Prosperity Sphere, Not Not, Charleston

Bars: Skylark, Rainbo, Inner town, Cole's, The Continental, Richards, Punch House,

Things can also take place at School or campus. Thanks Chat.

#### sulk embedding instructions:

put this at the end of the prompt:

[person in the room of sulk3:1]

#### download the embedding from

https://drive.google.com/file/d/1oO\_g3OmZrSiD5\_EXE88VT8d6K9niewM6/view?usp=share\_link

put the sulk3.bin into the "embeddings" folder in stable-diffusion-webui

#### Random example

((4 people at a party)), group photo, (gracerequejo :1.2) 1girl, and (Cassandrafitzgerald:1) 1boy, and (johncook:1.2) 1boy, and (yigitural:1.2) 1boy, beautiful, dancing, soft focus, close up, dark lighting, selfie, 8k photo

(gracerequejo:1) [SEP] (Cassandrafitzgerald:1) [SEP] (johncook:1) [SEP] (yigitural:1)

#### **RANDOMNESS PROMPTS:**

#### **RANDOMNESS:**

```
Prompt: UHD photo of (__people__:1.1), (__people__:1.1), (__people__:1.1), (__people__:1.1) at blah blah

ADetailer: ([__people__]:1.1) [SEP] ([__people__]:1.1) [SEP] ([__people__]:1.1)

([__people__]:1) [SEP] ([__people__]:1)
```

## MIXING REGIONAL PROMPTING AND DYNAMIC PROMPTING AND THE SULK EMBEDDING

Prompt:

```
UHD flash photo of 3people, (candid party photography) [person in the room of sulk3-1000] BREAK (__people__:1.1)BREAK (__people__:1.1)BREAK (__people__:1.1)
```

#### best settings ever

prompt

3people faces photo (candid party flash photography) multiple people in the white walls room [sulk3-1000] BREAK

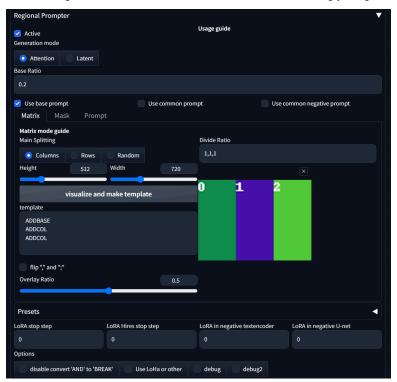
3 people, ( people ) BREAK

3 people, (\_\_people\_\_) BREAK

3 people, ( people )

#### negative:

(deformed iris, deformed pupils, semi-realistic, cgi, 3d, render, sketch, cartoon, drawing, anime, mutated hands and fingers:1.4), (deformed, distorted, disfigured:1.3), poorly drawn, bad anatomy, wrong anatomy, extra limb, missing limb, floating limbs, disconnected limbs, mutation, mutated, ugly, disgusting, amputation,





3people faces photo (candid party flash photography) multiple people in the white walls room [sulk3] BREAK

3 people, \_\_angle\_\_ (\_\_people\_\_) \_\_activites\_\_ BREAK

3 people, \_\_angle\_\_ (\_\_people\_\_) \_\_activites\_\_ BREAK

3 people, \_\_angle\_\_ (\_\_people\_\_) \_\_activites\_\_







